

The Logic of Comparative Theory Evaluation

1. Critique of Schattschneider's account
 2. Bayesian analysis of ad hocness
 3. The case of multiple predictions
 4. Quantitative predictions
 5. Bayesian Model for MSRP
 6. An alternative approach
 - The Methodological decision
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(2)

Bayes's Theorem

$$P(T/\bar{e}) = \frac{P(T/\bar{e}) \cdot P(e/T\bar{e})}{P(e/\bar{e})}$$

$$P(e/\bar{e}) = \sum_{i=1}^n P(T_i/\bar{e}) \cdot P(e/T_i\bar{e})$$

or

$$P(T/\bar{e}) = \frac{P(T/\bar{e}) \cdot P(e/T\bar{e})}{P(e/T\bar{e})P(T/\bar{e}) + P(e/\bar{T}\bar{e}) \cdot P(\bar{T}/\bar{e})}$$

(3)

Enhancement Ratio

$$\Omega = P(T|u \& e) / P(T|u)$$

Define $x = P(T|u)$
 $\varepsilon = P(e|u \& T \& u)$

then $\Omega = \frac{1}{x + \varepsilon(1-x)}$

Multiple Predictions

$$p_n = \frac{1}{1 - \varepsilon^n + \varepsilon^n/x}$$

$$p(e_{n+1}) = \varepsilon + \frac{1 - \varepsilon}{1 - \varepsilon^n + \varepsilon^n/x}$$

Simple model for ϵ

Theory describes M events, each with N possible results.

$n =$ Total number of theories

$m =$ number of theories which make a particular true prediction e

$$\text{then } \begin{cases} n = \{ f : \{M\} \rightarrow \{N\} \} = N^M \\ m = N^{M-1} \end{cases}$$

$$\text{Take } P(T_1|Z) = x, \quad P(T_i|Z) = \frac{1-x}{n-1}, \quad i \neq 1$$

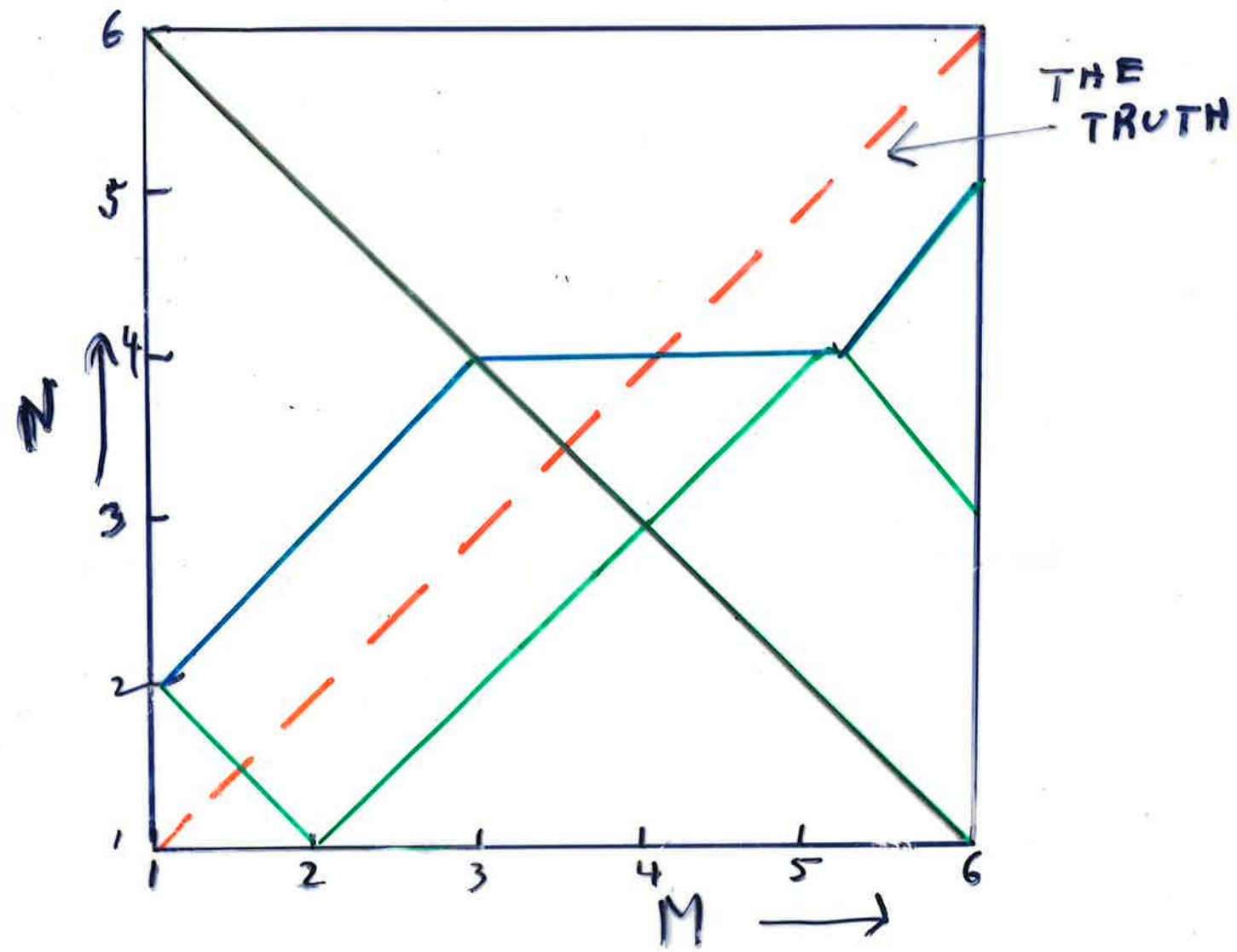
$$P(e|Z) = P(e|T_1, Z) P(T_1|Z) + \sum_{i \neq 1} P(e|T_i, Z) P(T_i|Z)$$

$$= x + (m-1) \cdot \frac{1-x}{n-1}$$

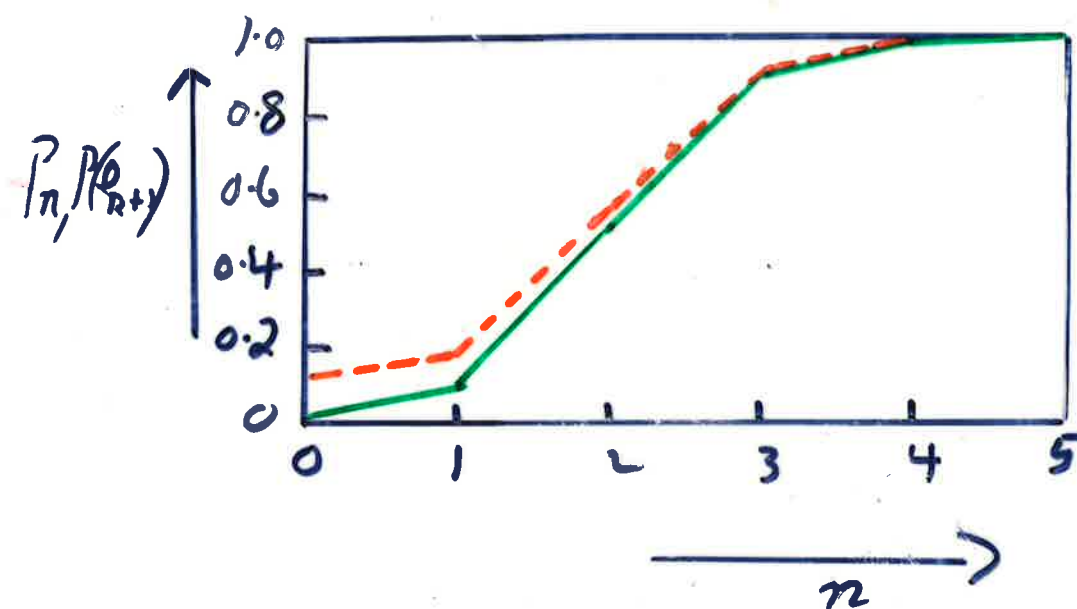
$$\text{Compare with } P(e|Z) = x + \epsilon(1-x)$$

$$\text{Hence } \epsilon = \frac{m-1}{n-1} = \frac{N^{M-1}-1}{N^M-1} \approx \frac{1}{N}$$

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Multiple predictions



— P_n
 - - P_{n+1}

$$\left. \begin{array}{l} \chi = 0.01 \\ \varepsilon = 0.1 \end{array} \right\}$$

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Dorning Analysis for Confirmation

Expand in Bayes' Theorem

$$P(e/c \& u) = P(e/c \& B \& u) \cdot P(B/c \& u) \\ + P(e/c \& \neg B \& u) \cdot P(\neg B/c \& u)$$

Write $P(c/u) = x$, $P(B/c \& u) = y$

Assume $P(e/c \& \neg B \& u) = P(e/\neg c \& u)$
 $= P(e/\neg c \& B \& u) = P(e/\neg B \& u)$
 $= A \text{ say.}$

Then $\Omega_c = \frac{y + A(1-y)}{xy + A(1-xy)}$

$\Omega_B = \frac{x + A(1-x)}{xy + A(1-xy)}$

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Dorling analysis for reputation

$$\lambda_C = \frac{1-y}{1-xy}$$

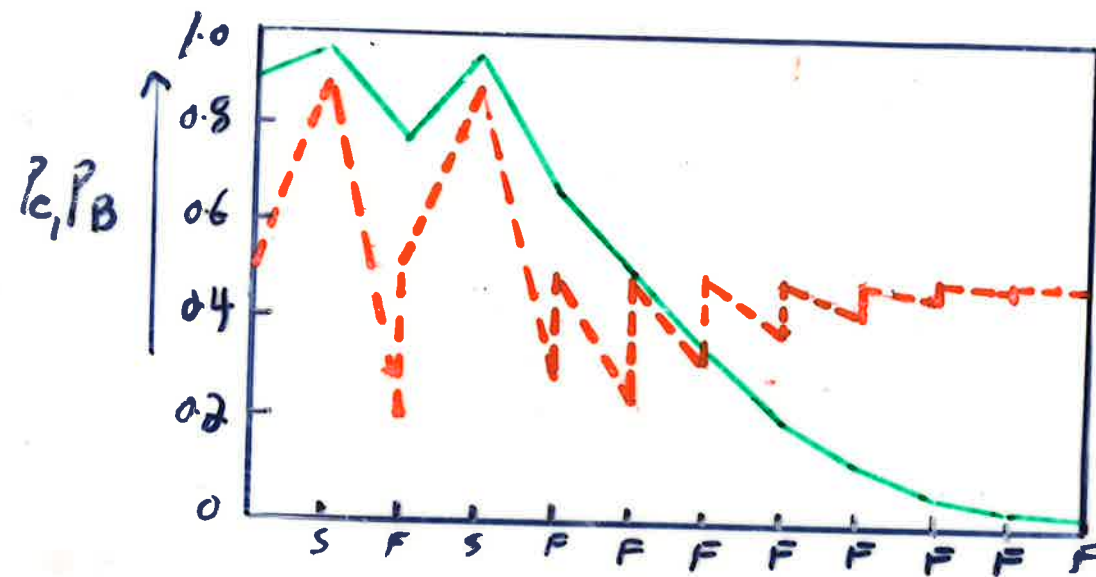
$$\lambda_B = \frac{1-x}{1-xy}$$

For asymmetric effect of reputation
take x close to unity, $y \ll \text{unity}$
i.e. $1-y \gg 1-x$.

The Dozling Model

$$x = 0.9, y = 0.5$$

$$h = 0.1$$



Sequence of predictions

— P_C
 - - - P_B

S = success
 F = failure

methodological decisions

Refutation

$$T = C \& B$$

$$P(T/\neg e) = P(C/\neg e) \times P(B/\neg e) = 0$$

choose $P(B/\neg e) = 0$

Confirmation

$$P(T/\neg e) = \Omega \times P(T/\neg e)$$

write $\Omega = \Omega_C \times \Omega_B$

where $P(C/\neg e) = \Omega_C \times P(C/\neg e)$

$$P(B/\neg e) = \Omega_B \times P(B/\neg e)$$

How do we choose Ω_C & Ω_B ?

(9)

The Factorization Constraints

$$\Lambda \cdot A \cdot B = (\Lambda_A \cdot A) \times (\Lambda_B \cdot B)$$

$$(1) \Lambda_A(1) = \Lambda_B(1) = 1$$

$$(2) \Lambda_A\left(\frac{1}{AB}\right) = \frac{1}{A}, \quad \Lambda_B\left(\frac{1}{AB}\right) = \frac{1}{B}$$

$$(3) \Lambda_A(\Lambda) \neq \Lambda_B(\Lambda) \text{ monotonic increasing functions of } \Lambda \text{ in range } (1, 1/AB)$$

$$(4) \Lambda_A \cdot A > \Lambda_B \cdot B \text{ for all } \Lambda \text{ in range } (1, 1/AB)$$

$$(5) \Lambda_A \cdot \Lambda_B = \Lambda$$

$$(6) \text{ Write } \Lambda_A(\Lambda) = f(A, B, \Lambda)$$

$$f(A, B, \Lambda' \Lambda'') = f(A, B, \Lambda') \times f\left[f(A, B, \Lambda') \cdot A, \frac{\Lambda}{f(A, B, \Lambda')} \cdot B, \Lambda''\right]$$

Solution satisfying conditions (1) - (6)

is
$$R_A = \frac{R(1-AB)}{1-A + A(1-B)R}$$

this does not satisfy condition

(7) $R_A = R_B = \sqrt{R}$ for $A=B$

Working soln for conditions (1) - (7) is

$$R_A = \frac{B + A(1-B)}{AB + A(1-AB)}$$

R is unique root in range $0 < R < 1$
of quadratic equation

$$R^2 \{ (1-A)(1-B) - (1-AB)^2 R \} + R \{ A+B - 2AB - 2AB(1-AB)R \} + AB(1-AB) = 0$$

The Memory Model

$$P_c(n_1, n_2, \dots, n_m) = P_c'(n_1, \dots, n_{m-1}) \times \Theta(\sqrt{x}, P_c'(n_1, \dots, n_{m-1}), \Omega^{(n_m)})$$

$$\text{where } P_c'(n_1, \dots, n_{m-1}) = F(m-1) \times P_c(n_1, \dots, n_{m-1})$$

$$\text{and } \Omega^{(n_m)} = \Phi(\varepsilon, n_m, \sqrt{x} \cdot P_c'(n_1, \dots, n_{m-1}))$$

$$\Omega^{(n_1)} = \bar{\Phi}(\varepsilon, n_1, x)$$

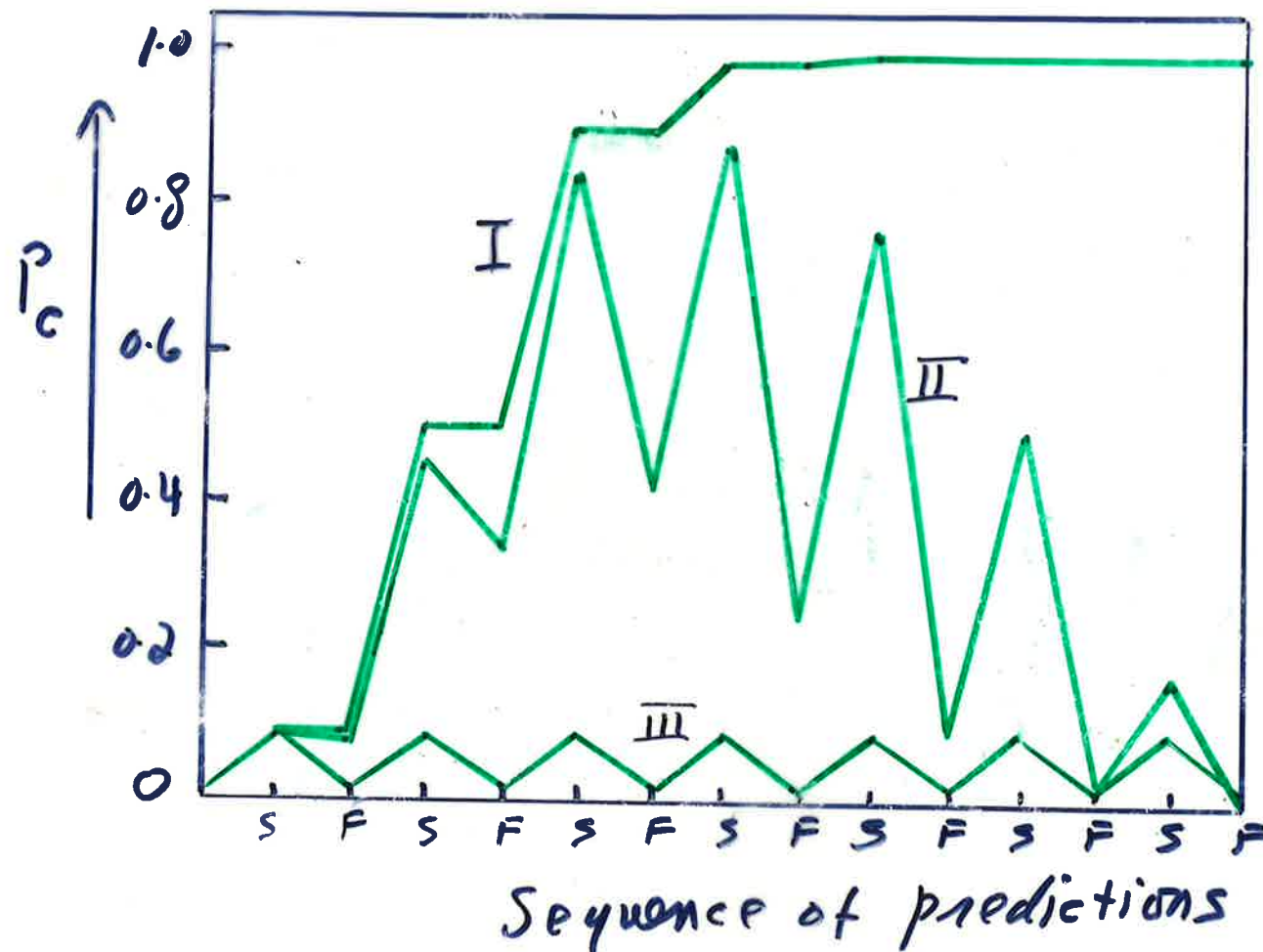
$$\text{with } P_c(n_1) = \sqrt{x} \Theta(\sqrt{x}, \sqrt{x}, \Omega^{(n_1)})$$

$$\text{and } \Theta(A, B, C) = \frac{C(1-AB)}{1-B+B(1-A)C}$$

$$\bar{\Phi}(A, B, C) = (C(1-AB) + A^B)^{-1}$$

$$F(m) = (1 + \exp(m-m_0))^{-1}$$

• The Life & Death of a Research Programme



- I Pure Lakatos
 II Pure Lakatos plus memory
 III Pure Popper

$$\left. \begin{array}{l} \sqrt{x} = 0.01 \\ \epsilon = 0.1 \\ m_0 = 3 \end{array} \right\}$$